

US EPA ARCHIVE DOCUMENT

# Evaluating Voluntary Programs in the United States+

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# Voluntary Climate Programs: Climate Wise

- Established 1993; continued until 2000. Focus on non-utility industrial sector.
- Required baseline emission estimate (but not inventory).
- Required to identify mitigation actions, goal for 2000.
- Report activities via 1605(b).
- Gained technical assistance; annual workshop.

# Voluntary Climate Programs: 1605(b)

- Established under EPACT 1992; began 1994.
- Required reporting of emission reductions, with flexibility over
  - Whether entity or project
  - Reference year or hypothetical reference
  - Absolute or intensity reductions
- Open to any individual or business; dominated by electric power industry
- Important benefits (EIA, 2002):
  - Teach corporations how to estimate emissions and mitigation options
  - Sharing experience concerning mitigation activities
  - Evidence for evaluating other voluntary programs
  - Illuminate accounting issues related to future emission regulation

# Participation in Programs (raw / not linked to LRD)

Join year	1605(b)	ClimateWise
1994	43	8
1995	105	37
1996	37	179
1997	26	138
1998	17	106
1999	61	89
2000	35	144
2001	59	
Subtotal	383	671

# Key Challenges

- Measuring outcome
  - Need data on emission outcomes before and after policy, for both participants and non-participants.
- Addressing selection
  - Participants and non-participants may not look the same and/or participation may depend on various characteristics unrelated to the program but correlated with outcome.

# Proposed Solutions

- Use Census data on energy use (expenditures on fuel and electricity) to proxy for emissions. Available for both participants and non-participants; requires working at Census Bureau to access confidential data and link to participation information.
- Address selection through two alternative models.

# Selection Problem and Solutions

$$Y_{i,t} = f(X_{i,t}) + g(X_{i,t})D_{i,t} + u_{i,t}$$

- $Y$  is emissions / energy use;  $D$  is participation;  $X$  are covariates (location, industry, size).
- $g(X_{i,t})$  measures program effect on outcome.
- Potential problems
  1.  $\mu$  correlated with  $D$ .
  2. miss-specification of  $f$  and  $g$ .
- Solutions
  - Structural model of selection and correlation with  $\mu$  (Heckman-Hotz). *Requires excluded variable predicting selection and not outcome.*
  - Propensity score matching.



# Heckman-Hotz

- Consider joint estimation of selection model and outcome model:

$$D_{i,t}^* = \delta \cdot Z_{i,t} + v_{i,t}$$
$$Y_{i,t} = f(X_{i,t}) + g(X_{i,t}) D_{i,t} + u_{i,t}$$

- Here,  $(u_{i,t}, v_{i,t})$  are jointly normal,  $Z_{i,t}$  includes at least one additional variable than  $X_{i,t}$ , and  $D_{i,t}^*$  is a continuous latent variable, with  $D_{i,t} = 1$  when  $D_{i,t}^* > 0$ .

- Estimate selection model using probit; insert additional regressor in outcome model,

$$E[u_{i,t} | v_{i,t}] = \lambda(D_{i,t}, Z_{i,t})$$

# Problems with Heckman-Hotz

**Table 1: EPA Climate Wise program, effect of program on logged cost of electricity after 2 years, Heckman-Hotz approach**

Cohort	w/o correction	with correction	sample	participants
1994	0.06 (0.02)*	0.60 (0.09)*	19627	809
1995	0.04 (0.04)	-0.16 (0.14)	34880	335
1996	0.02 (0.03)	0.36 (0.21)	31253	656
1997	-0.02 (0.03)	-0.29 (0.18)	17534	835
1998	0.01 (0.02)	-0.75 (0.16)*	30693	1063
1999	0.05 (0.12)	-1.42 (0.71)*	33971	96

- Program effects on energy costs are  $\pm 100\%$ .
- *Excluded variables (membership in advocacy organization & distance to EPA regional office) are not effective at predicting participation.*

# Propensity Score Matching

- Estimate participation model and predict propensity to join for each plant in each year.
- Consider each participating plant; find non-participating plant with closest propensity value (nearest neighbor) in the join year.
- Sample without replacement.
- Estimate separate selection model for each horizon (1, 2, and 3 years) where program effects are computed.

# Participation Model

probability of joining in year  $t$   
(assuming plant  $i$  has not yet joined)  $= h(t) \exp \left( \begin{aligned} &\beta_{size} \ln TVS_{i,t-1} + \beta_{elec} \ln EE_{i,t-1} + \beta_{fuels} \ln CF_{i,t-1} \\ &+ [\text{all quadratic combinations of size, elec, fuels}] \\ &+ \beta_{growth} (\ln TVS_{i,t+h} - \ln TVS_{i,t-1}) \\ &+ \sum_{\text{industries } j} \beta_j 1(M_i = j) + \sum_{\text{region } k} \beta_k 1(G_i = k) \end{aligned} \right)$

- Cox proportional hazard model of probability of plant  $i$  choosing to join in year  $t$ .
- includes lagged total value of shipments (TVS), electricity expenditures (EE), cost of fuels (CF), plus linear and quadratic terms, interactions
- future growth rate in shipments (  $h = 1$ -,  $2$ -, or  $3$ - year lead vs.  $1$  year lag).
- includes census region  $G$  (9 values) and industry  $M$  (2-digit) dummy variables.

# Model of Program Effects Using Pairwise Matched Participants / Controls

$$\Delta Y_s = \frac{\sum_{i,t} \left( Y_{\text{participant } i, t+s} - Y_{\text{participant } i, t-1} \right) - \left( Y_{\text{control } i, t+s} - Y_{\text{control } i, t-1} \right)}{\sum_{i,t} 1 \left( \begin{array}{l} \text{participant } i \text{ joined in } t \\ \text{and observed in } t+s \text{ and } t-1 \end{array} \right)}$$

- $Y_{d,i,t}$  is the relevant variable (total value of shipments, fuel and electricity expenditure) in pair  $i$  at time  $t$ .
- $\Delta Y_s$  is the average program effect after  $s$  years relative to the year before the joinyear (for output, fuel and electricity expenditures)

# Propensity Score Results

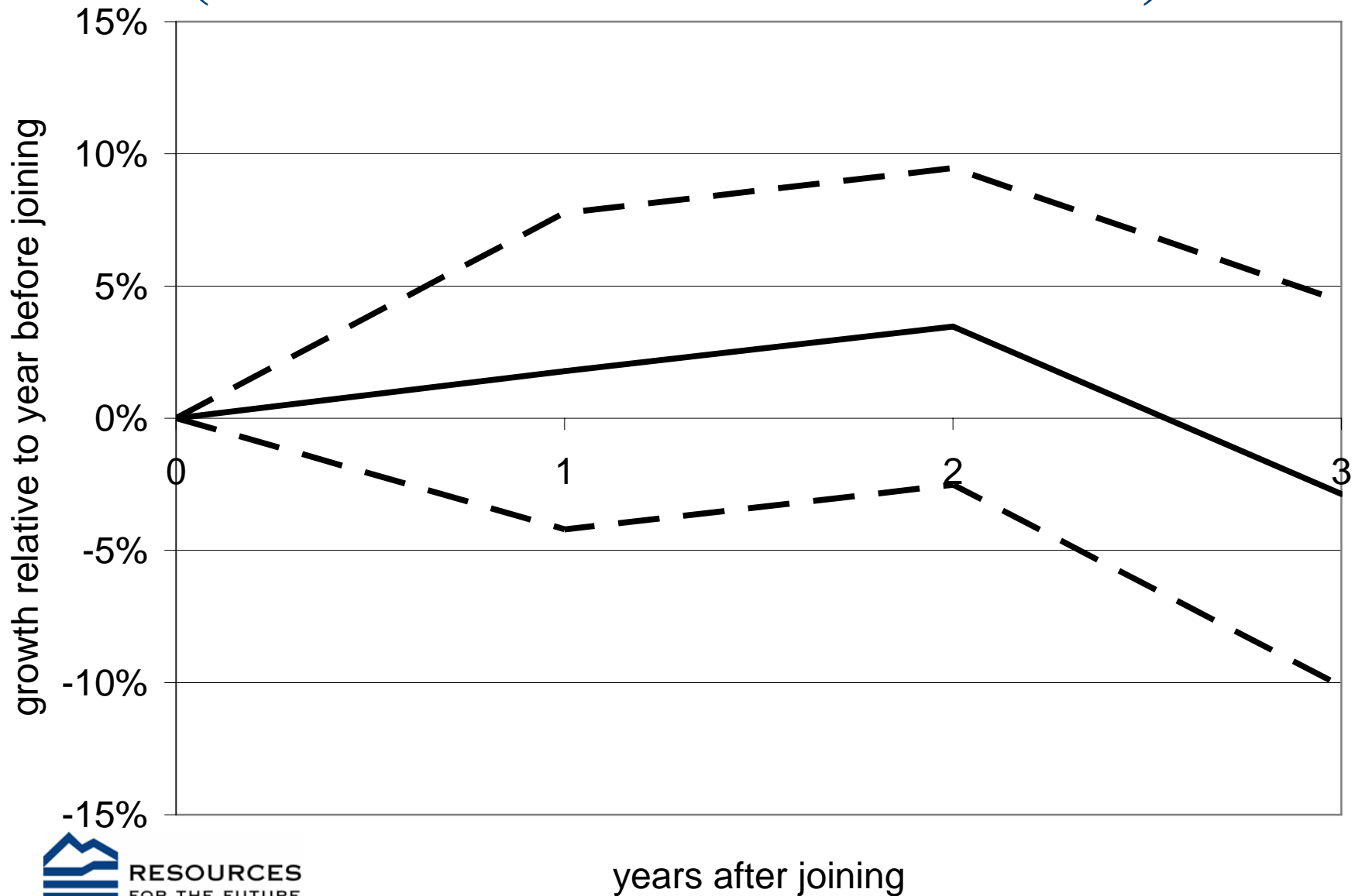
## (median estimates with all controls)

*Effect of program participation on energy expenditures  
(fractional change)*

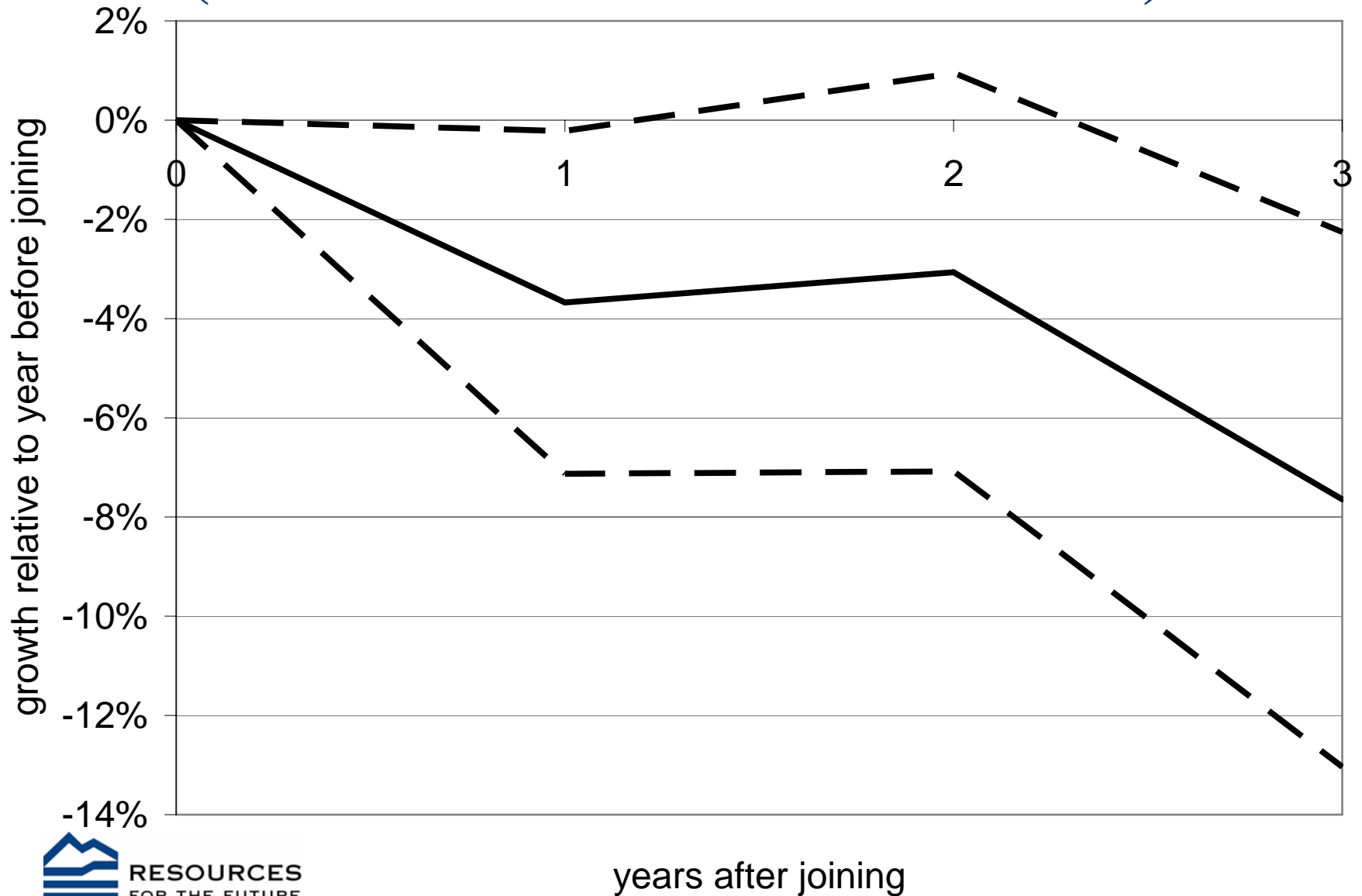
	1605(b)		ClimateWise	
	Fuel	Electricity	Fuel	Electricity
1-year effect	0.02 (0.03)	-0.04* (0.02)	-0.01 (0.03)	0.00 (0.01)
2-year effect	0.03 (0.03)	-0.03 (0.02)	0.03 (0.03)	0.02 (0.02)
3-year effect	-0.05 (0.06)	-0.05* (0.03)	-0.01 (0.03)	-0.01 (0.02)

- Effects are no more than 5% with most general specification. Zero for ClimateWise.
- Other specifications lead to a wider range of median estimates from –8% to +5% (positive effect is transitory)

# 1605(b) Effect on Fuel (with 95% confidence interval)

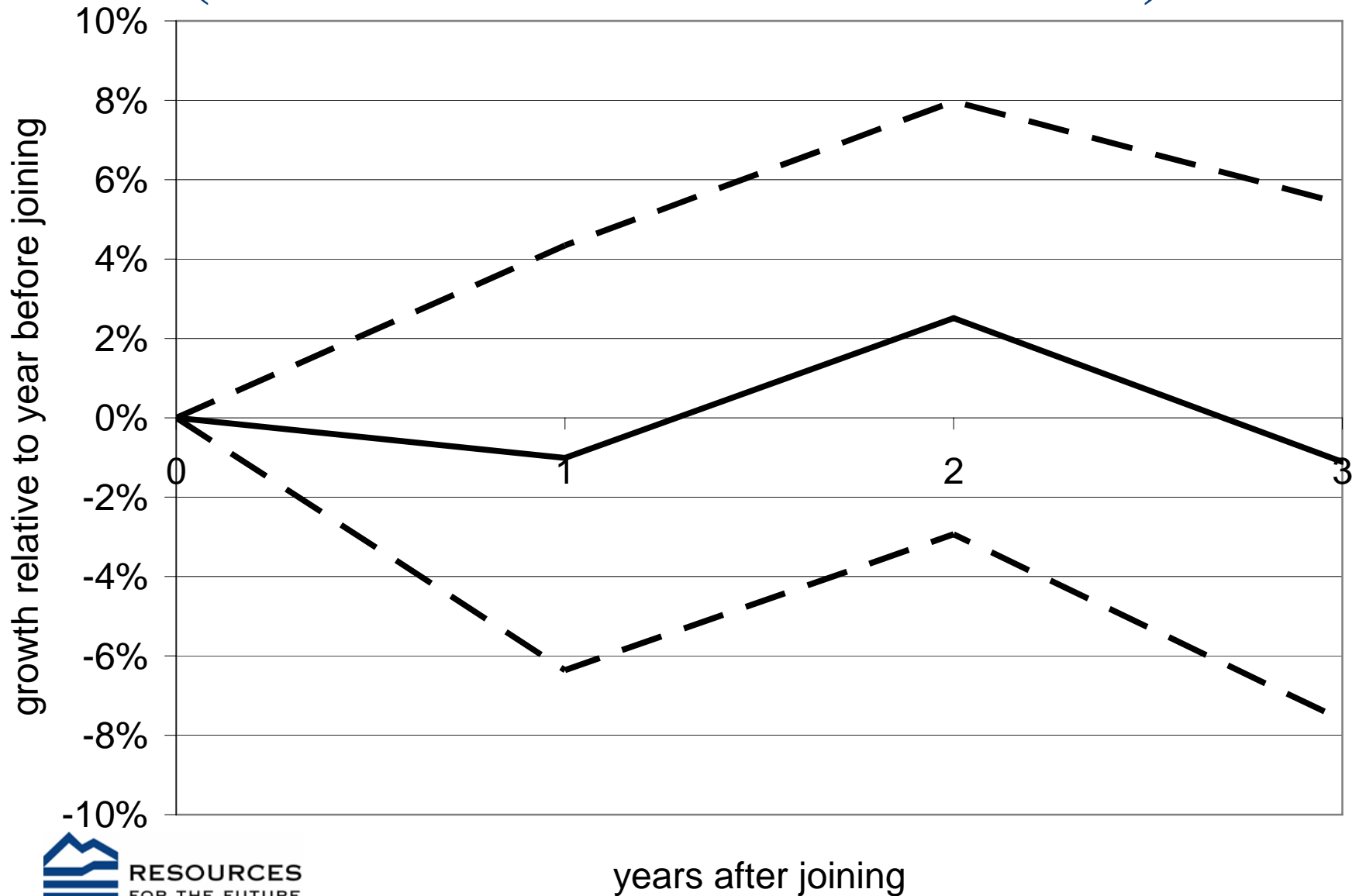


# 1605(b) Effect on Electricity (with 95% confidence interval)

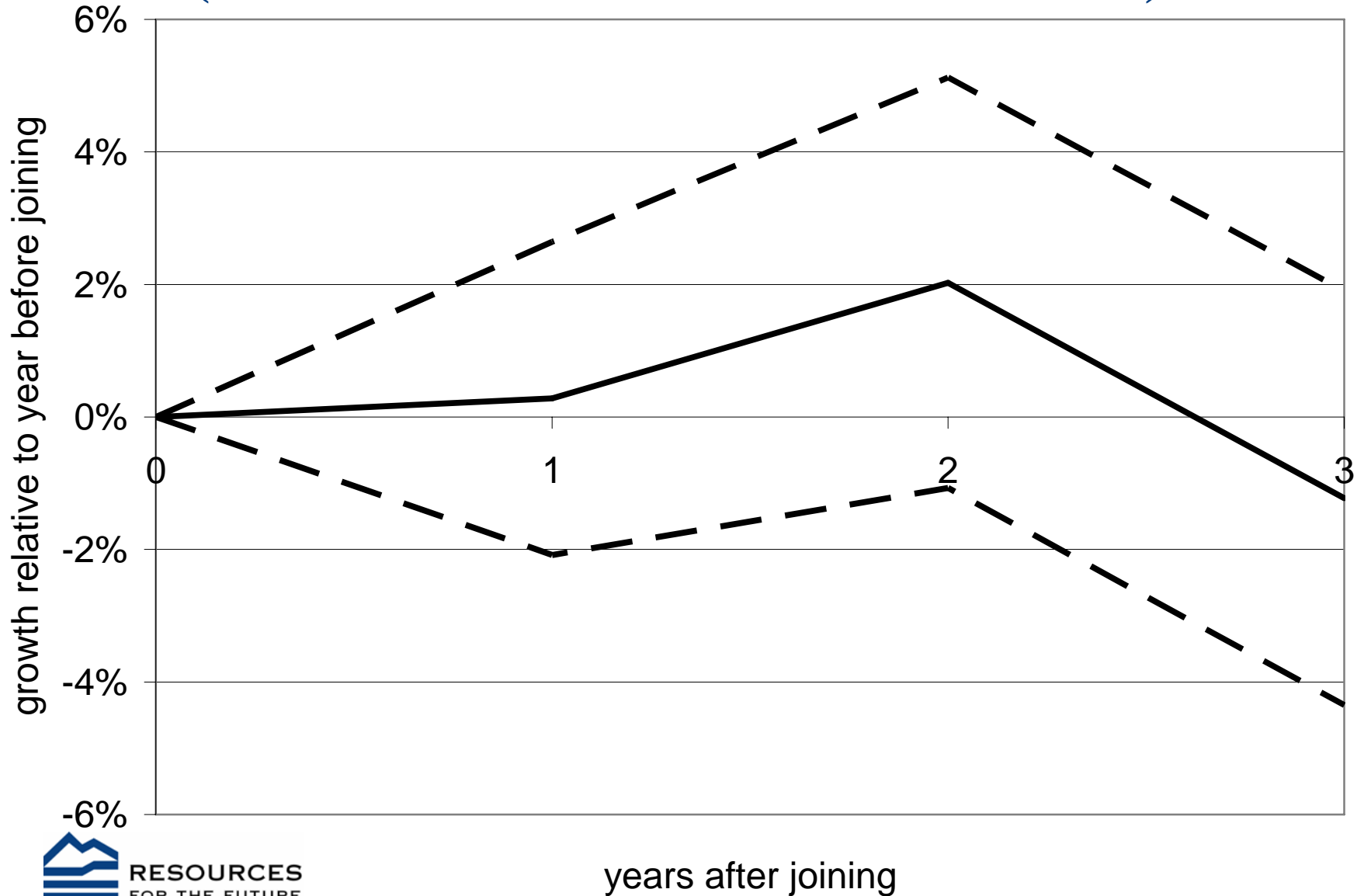




# ClimateWise Effect on Fuel (with 95% confidence interval)

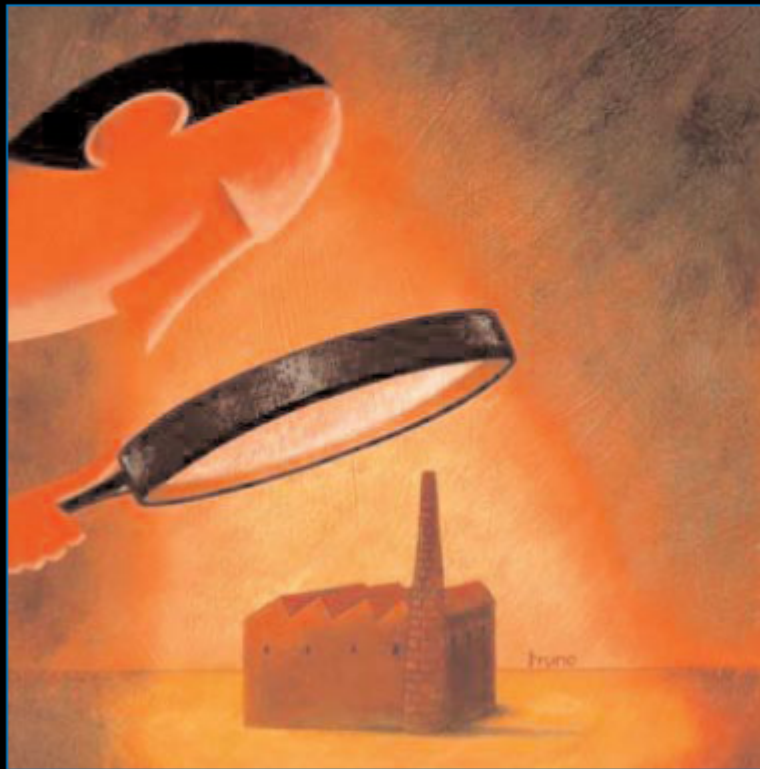


# ClimateWise Effect on Electricity (with 95% confidence interval)



# Reality Check

**The Nature and Performance of  
Voluntary Environmental Programs in  
the United States, Europe, and Japan**



**Richard D. Morgenstern and  
William A. Pizer, editors**



# What are “voluntary programs”?

## Why do we care?

- Types of programs
  - Unilateral agreements
  - Public voluntary programs
  - Negotiated agreements
- Varied and expanding use
  - 87 EPA programs, 1.6% operating budget
  - Dozens more in states, other federal agencies
  - Hundreds of VP/VAs in Europe
  - Thousands in Japan
- BUT, do voluntary programs deliver significant environmental gains relative to a realistic baseline, i.e., do they change behavior?
  - If so, how large are the gains?
  - Do results differ for toxics vs energy programs?
  - What else affects program impact?

# Motivation

- Business
  - Get ‘hands on’ experience
  - Enhance reputation with customers, government, investors, communities, etc.
  - Benefit from government-provided technical assistance.
  - Help shape future requirements; improve relationship with regulators
- Government
  - Get ‘hands on’ experience in the absence of regulatory mandate
  - Experiment with more holistic approaches vs traditional regulation
  - Build public support for future action
  - Build bridges to industry, e.g., via technical assistance
- Environmental groups (mixed reaction)
  - Some applaud VP’s as means to build support in public, industry
  - Some fear regulatory capture, distraction from real work of environmental protection, shift in focus from worst polluters to more progressive firms

# Table 1-1: Selected Characteristics of Case Studies

<b>Program</b>	<b>Author(s)</b>	<b>Years of Operation</b>	<b>Energy, CO2 (GHGs), or Toxics</b>	<b>Industry or Household</b>	<b>Program Type</b>
<b>33/50 (US)</b>	<b>Khanna</b>	<b>1991-1996</b>	<b>Toxics</b>	<b>Industry</b>	<b>Public Voluntary Program</b>
<b>Japanese Keidanren</b>	<b>Wakabayashi and Sugiyama</b>	<b>1997-</b>	<b>CO2</b>	<b>Industry</b>	<b>Negotiated agreement</b>
<b>UK Climate Change Agreements</b>	<b>Glachant and Muizon</b>	<b>2001-</b>	<b>CO2</b>	<b>Industry</b>	<b>Negotiated agreement</b>
<b>Danish Energy Efficiency Agreements</b>	<b>Krarup and Millock</b>	<b>1996-</b>	<b>CO2</b>	<b>Industry</b>	<b>Negotiated agreement</b>
<b>German Cement Industry</b>	<b>Bohringer and Frondel</b>	<b>1995</b>	<b>CO2</b>	<b>Industry</b>	<b>Unilateral agreement</b>
<b>Climate Wise (US)</b>	<b>Morgenstern, Pizer and Shih</b>	<b>1993-2000</b>	<b>GHGs</b>	<b>Industry</b>	<b>Public Voluntary Program</b>
<b>California Demand Side Management</b>	<b>Sanstad</b>	<b>Early-mid 1990s</b>	<b>Energy</b>	<b>Household</b>	<b>Public Voluntary Program</b>

# 33/50 Program

- Followed development of TRI
- Focus on measurable reductions (33%, 50%) for 17 TRI chemicals in major industries (1991)
- Actual reductions clearly exceeded goals
- Sophisticated studies find program reduced emissions, controlling for self-selection, especially for larger firms
- Partly attributable to fear of regulations
- Some evidence suggests no/negative gains beyond Montreal Protocol substances

# Keidanren Voluntary Action Plan

- Involves large firms representing 80% of industrial, electric emissions (almost half of Japan's total emissions) (1997)
- Targets negotiated for sectors, not firms
- So far, emissions below target levels
- Reductions attributed to industry, gov't cooperation, fear of regulation, firms' social awareness
- Questions about BAU estimates, stringency of goals
- Is program really voluntary?



# UK Climate Change Agreements

- CCAs part of tax (\$9-18/ton of CO<sub>2</sub>), and emissions trading policies (2001)
- Intensity or fixed targets negotiated with gov't
- Covering 12,000 sites = 44% UK emissions
- 80% rebates of levy for meeting CCA goals
- Goals exceeded (based on observed permit prices), although stringency in question
- Overall, authors find that CCAs make small contribution

# Denmark's Energy Efficiency Agreements

- VAs part of policy package involving CO<sub>2</sub> taxes (\$18/ton) on industry (1996)
- Negotiated agreements based on audits, adoption of energy efficiency measures. No quantitative targets
- 100% tax rebates for participants
- Audit eventually dropped
- Using data from 60 firms, authors find some reductions in early years, although quite modest reductions overall

# German Cement Industry

- Unilateral commitment by major sectors (not firms) for 20% cuts below 1987 levels by 2005; case focuses on cement industry (1995)
- By 2000 most goals met; target raised to 28% reduction
- Trend regression used to establish baseline using historical data
- Actual emissions same as forecast BAU (+/- 5%)
- Authors recommend firm specific targets; negotiated instead of unilaterally set

# Climate Wise

- EPA program involving negotiated agreements with 600+ firms (1993)
- Emissions based program; TA, other incentives offered for joining
- Comparisons with matched set of non-participants used to determine what would have happened anyway
- Authors find modest differences in fuel (-) and electricity (+) use in early years; no significant differences later on

# Residential DSM in California

- Utilities started providing free technical information to single family houses in 1970s
- Two of three evaluations indicate savings ‘that would not have occurred without programs’
- One study finds changed maintenance and other practices more important than use of new equipment
- Some evidence that provision of information by authoritative source is key

**Table 9-1: Quantitative comparison of the effect of voluntary programs on behavior**

	<b>Quantity measured</b>	<b>Estimated Effect</b>	<b>Scope</b>	<b>Baseline</b>	<b>Comment</b>
33/50 Program	Aggregate toxic releases	28%	Participating chemicals facilities	Non-participants with self-selection model	Effect reversed when ODS excluded.
UK Climate Agreements	GHG emissions	9%	Participating industries	Negotiated forecast	Baseline criticized; considerable over-achievement.
Danish Energy Efficiency Agreements	Energy Use	4-8%	Participating facilities	Non-participants	Estimate based on 60 participants.
German Cement Industry GWP Declaration	Energy per unit of cement	0	German cement industry	Econometric forecast using historic performance	Baseline error band is +/- 5%. 2005 target achieved by 2000.
Japanese Keidanren	CO <sub>2</sub> emissions	5%	Participating industries	Keidanren forecast of 2010 BAU	Basis of BAU estimate unclear.
Climate Wise	Fossil energy expenditures	3%	Participating facilities	Matched non-participants	Electricity expenditures estimated to rise 6%. Margin of error is +/- 5% and both effects vanish after 1-2 years.
California Demand Side Management	Natural gas & electricity demand	2-4%	Participating households	Non-participants	Covers three programs; some evaluations more carefully matched non-participants / controlled for self-selection

# Conclusions

- Hard to reject conclusion of 5% reduction for energy programs, +/- 5%. Thus, evidence that VPs do change behavior, but not suitable for major reductions
- Significant differences exist between energy and toxics, although clear limitation on toxics as well
- Incentives have modest impact on reductions achieved among participants, potentially larger impact on level of participation
- Efforts to increase program breadth (i.e., many participants) may yield greater environmental gain than efforts to increase depth (big cuts in emissions for individual firms) (broad vs deep)
- More attention needed on baselines for evaluation, including both forecasts and control group approaches
- Subtle changes in social attitudes and corporate practices may be significant but are difficult to measure